



# Next Level Python

Tips, tool, and techniques for being  
highly productive in Python

# Topics

Prelude

Data structures

Iterators

Functions, Files, & Modules

**Exercises**

Using an IDE

Debugging

Conda Environments

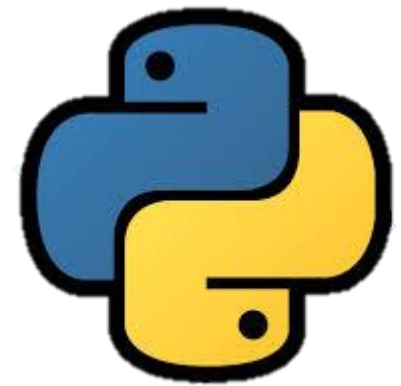
**Exercises**

Code snippets, if you want to follow along:

[bit.ly/2ScFewN](https://bit.ly/2ScFewN)

# Topic 0

# **Prelude**



## Prelude

# What is Python?

- Invented in the early '90s by Guido Van Rossum

~~BDFL~~



- **CPython Interpreter**
  - Reads your code 📖
  - Parses it 👁️
  - Checks for **syntax** errors ✓
  - Generates **instructions** 🖱️
  - Steps through instructions ▶️

# Prelude

## 2 vs. 3

Python 3 > Python 2

Stick to 3.5+



Python 3



Python 2

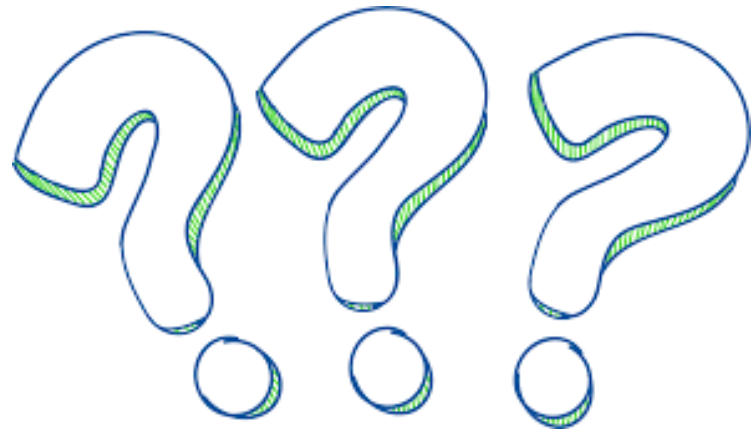
# Prelude

## Variables

- Sounds simple, but worth clarifying

3      a = 1

4      a = 2



What just happened?

Did we change 1? Or a?

What happened to 1?

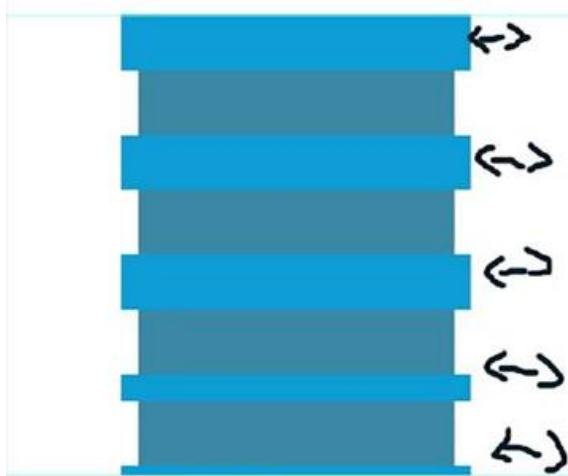
# Prelude

## Variables

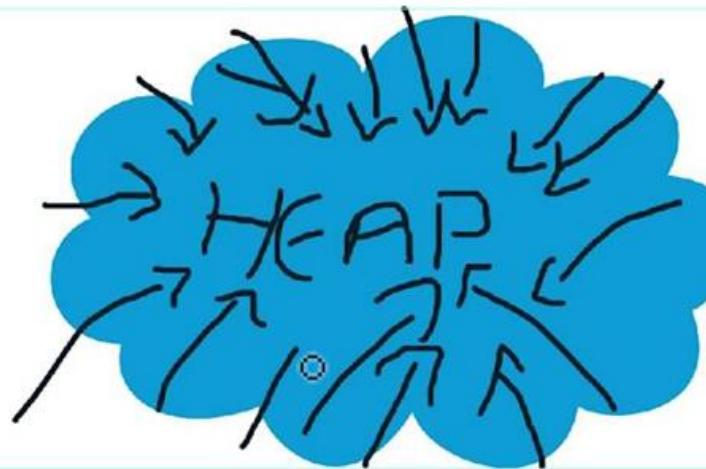
- Values are **objects**
  - **objects** live on the heap
- Names tell you how to find an object
  - **names** live on the heap

3	a = 1
4	a = 2

### Stack



### Heap





# Prelude

## Variables

```
3    foo = [3,1,2]
4    bar = sort(foo)
5    print(foo)
6    # [1,2,3]
```

What happened to foo?

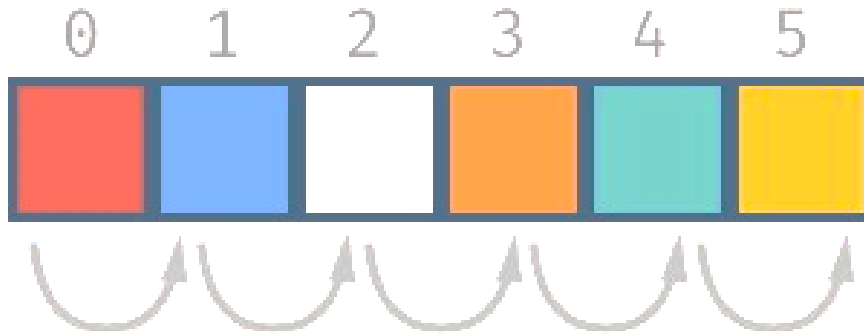
## Prelude

# Beyond CPython

- There are other ways of running your Python code:



The best way to get CPython + Scientific libraries



Topic 1

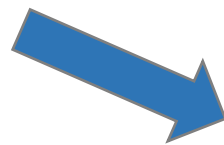
# Data Structures

## Topic 1

# Data Structures

### Lists

- Most of us are familiar
- Store data in **order**
- Can **loop** through
- Can look up by **index**



```
abc = [1,2,3]  
print(abc[2])  # 3
```

- Slow **searching**

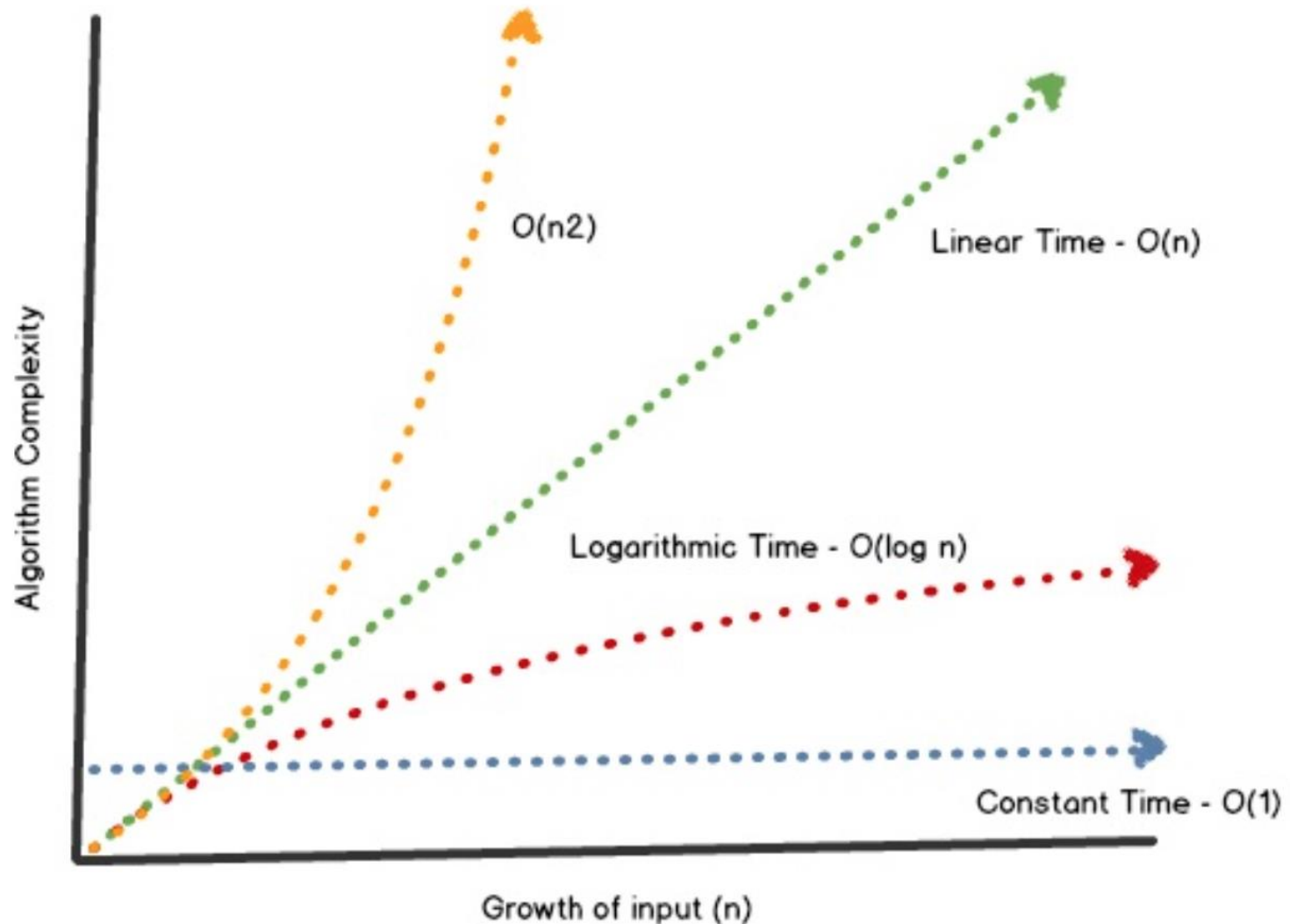
# Data Structures

## Dictionaries

- Store data as key: value pairs
  - Keys can be:
    - Numbers
    - Strings
    - Tuples of numbers, strings
  - Values can be: anything!
  - “**Instant**” lookup by key!
    - $O(1)$  even if  $1e6$  entries!
  - Slower than lists for looping

# Data Structures

## Dictionaries



# Data Structures

## Dictionaries

- Use cases:
  - Store properties of something in one place

- Example:

```
my_particle = {  
    'x': 12.0,  
    'y': 13.5,  
    'z': -12.0,  
}
```

```
def print_particle(particle_dict):  
    print("x: ", particle_dict['x'])  
    print("y: ", particle_dict['y'])  
    print("z: ", particle_dict['z'])
```

# Data Structures

## Dictionaries

- Use cases:
  - Data you have to access by a name or ID
  - Catalogs
  - Filenames for datasets

- Example:

```
30 runs_data_number = {
31     'M29-768': 345
32     'M35-1536': 138
33     ...
34 }
35
36 print(runs_data_number['M35-1536']) # 138
37
38 for run in dataset:
39     run_number = runs_data_number[run]
40     file = open(run + '/' + data_number, 'r')
41     ...
```



# Data Structures

## Dictionaries – Trick!

- You can use tuples as dictionary keys
- Use this as a sparse matrix, esp. for non-numerical data!

```
17  # Dictionaries as matrix/table
18  mat = {}
19  mat[(1, 9)] = 5
20  mat[(23000, 1e12)] = 4
21  for (keya, keyb) in mat.keys():
22      val = mat[(keya, keyb)]
23      print(f"a: {keya}  b: {keyb}  val: {val}")
24  # Output:
25  # a: 1  b: 9  val: 5
26  # a: 23000  b: 10000000000000.0  val: 4
```


- This would require on the order of 2TB of RAM if you used a list!
- Great way to store values that belong at the intersection of two other things, e.g. (computer A, and computer B) have an open connection C

# Data Structures

## Dictionaries – Trick!

- You can use dictionaries to hold arguments for functions, apply them later

```
111 # Dictionaries: Tip for Function arguments
112 def myfunc(min, max, step):
113     pass
114
115 kwargs = {"min": 0, "max": 10, "step": 14}
116 myfunc(**kwargs)
117
118 all_the_args = [
119     {"min": 0, "max": 10, "step": 3},
120     {"min": 0, "max": 12, "step": 4},
121     {"min": 6, "max": 10, "step": 1}
122 ]
123 for kwargs in all_the_args:
124     myfunc(**kwargs)
125
```



# Data Structures

## Dictionaries – Tip!

- You can use lists, and other dictionaries \*inside\* dictionary values!

```
28 # Dictionaries with lists for values
29 # Table of companions for some galaxy
30 galaxies = ["M31", "MW"]
31
32 companions = {}
33 companions["M31"] = ["M32", "M110", "etc"]
34 companions["MW"] = ["CMd", "LMC", "SMC", "etc"]
35
36 for galaxy in galaxies:
37     cs = companions[galaxy]
38     print(f"{galaxy} has {len(cs)} companions!")
39 # M31 has 3 companions!
40 # MW has 4 companions!
```

# Data Structures

## Dictionaries – Technique!

- If you're calling some function, e.g. from matplotlib lots of times with slight differences, use a dictionary to hold defaults

```
130 my_defaults = {  
131     "marker": "o",  
132     "color": "blue",  
133     "linewidth": 3,  
134     "alpha": 0.6  
135 }  
136  
137 plot([1, 2, 3], **my_defaults)  
138 plot([3, 4, 5], **my_defaults)
```

Better yet! Make a function!

# Data Structures

## Sets

- Sets are like a “**bag**” of values
- Values can be strings, numbers, tuples, etc.
- Values are unique.
  - Adding the same thing twice has no effect
- Not ordered
- “**instant**” adding and removing values
- “**instant**” membership testing
- Operations for
  - Testing
  - Grouping
  - Differencing
  - etc.

# Data Structures

## Sets

- Use case:
  - You have a data set, with some possibly overlapping groups
  - You want to focus on one group at a time
  - Solution: Put the group in a set
  - You can easily isolate just the data in the group

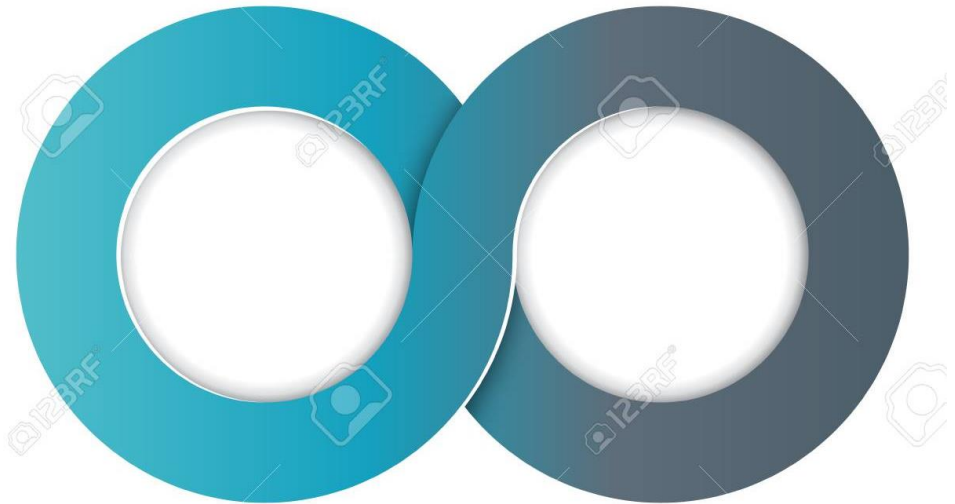
# Data Structures

## Sets

```
15 bar = set() # or {},}
16 bar.add(1)
17 bar.add(2)
18
19 print(2 in bar) # True
20 print(3 in bar) # False
21
22 baz = set([2,3,4])
23 print(bar & baz) # {2} in both
24 print(bar | baz) # {1,2,3,4} in either
25 print(bar ^ baz) # {1,3,4} in either, but not both!
26 print(bar > baz) # False is everythin in bar, also in baz?
```

Topic 2

# Iterators





## Topic 2

# Iterators

- Generators create “lazy sequences”, called iterators

```
60     foo = range(0, 10000000000000, 1)
61     for b in foo:
62         print(b)
63         if b > 10:
64             break
```

This **works!**

Why doesn't my computer crash?

# Iterators

## Behind the scenes

- Behind the scenes, they are a function that returns more than one value.
  - Every time you run the function, you get the next value!

### Definition:

```
66 def count_down(stop):
67     value = 0
68     while value > stop:
69         yield value
70         value -= 1
```

### Usage:

```
72 for i in count_down(-5):
73     print(i, end=', ')
74 print()
75 # 0, -1, -2, -3, -4
```

# Iterators

## Useful Iterator Functions

- enumerate
  - Loop through things with indexes

```
78  # Iterators: enumerate
79  letters = ['a', 'b', 'c', 'd', 'e']
80
```

# Iterators

## Useful Iterator Functions

- zip
  - Group items from multiple lists into tuples

```
letters = ['a', 'b', 'c']
```

```
numbers = [5, 4, 3]
```

```
list(zip(letters, numbers))
```

```
# [('a', 5), ('b', 4), ('c', 3)]
```

```
for letter, number in zip(letters, numbers):
```

```
|     print(letter + str(number))
```

```
# a5
```

```
# b4
```

```
# c3
```

# Iterators

## Useful Iterator Functions

- zip
  - Trick! You can *unzip* as well, using \*

```
107     # Zip trick
108     print(zipped)           # [('a', 5), ('b', 4), ('c', 3)]
109     print(list(
110         |     zip(*zipped)   # (['a', 'b', 'c'], [1,2,3])
111         ))
```

# Iterators

## Useful Iterator Functions

- map, filter, etc
  - Really useful for lists, but if using numpy there are better better ways

# Iterators

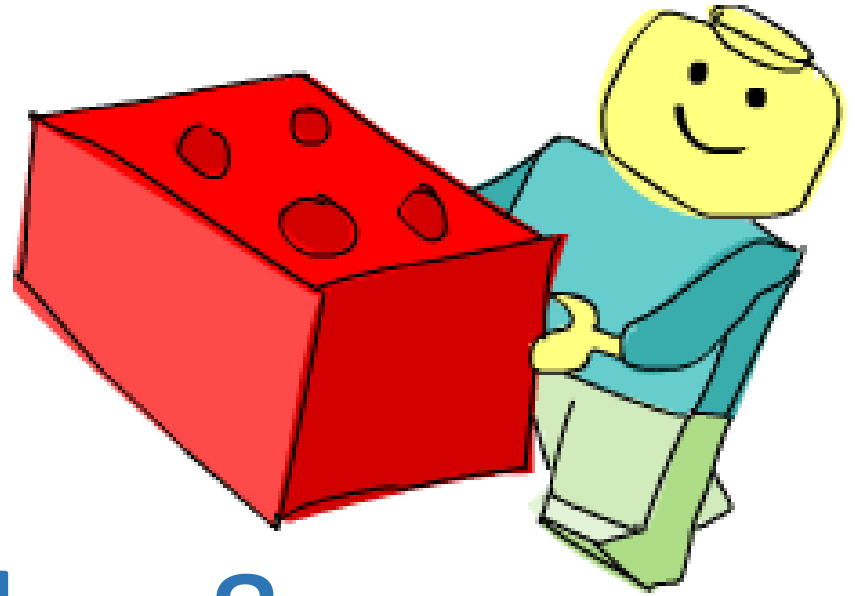
## Generator Expressions

- Generator comprehensions by example

```
8   # Make a sequence from 0 to 10^18
9   gen1 = range(1, 1000000000000000000)
10  # Square them all
11  gen2 = (i ** 2 for i in gen1)
12  # Include only odd numbers
13  gen3 = (i for i in gen2 if i % 2 == 0)
14  # Enumerate them all
15  gen4 = enumerate(gen3)
16
17  for i, value in gen4:
18      print(i, value)
19      if value > 120:
20          break
```

Prints:

0	4
1	16
2	36
3	64
4	100
5	144



Topic 3

# Functions, Files & Modules

Oh, and packages too



# Topic 3

## Functions

- **DRY: Do not Repeat Yourself**
  - Use functions to pull out repeated code, patterns
- **Safety**
  - Variables inside functions can't **leak** out
  - Easier variable naming
- **Abstraction**
  - Once you've written one type of data analysis, don't need to worry about the details again

The diagram illustrates the components of a Python function definition. It shows the code: `def fahr_to_celsius(temp):` on the first line and `return ((temp - 32) * (5/9))` on the second line. Arrows point from labels to specific parts of the code: 'def statement' points to 'def', 'name' points to 'fahr\_to\_celsius', 'parameter names' points to '(temp)', 'body' points to the entire second line, 'return statement' points to 'return', and 'return value' points to the expression '((temp - 32) \* (5/9))'.

```
def fahr_to_celsius(temp):  
    return ((temp - 32) * (5/9))
```

Labels and arrows:

- def statement → `def`
- name → `fahr_to_celsius`
- parameter names → `(temp)`
- body → `return ((temp - 32) * (5/9))`
- return statement → `return`
- return value → `((temp - 32) * (5/9))`

# Functions

## Rules of Thumb

### Create a function if:

- You write the same thing 2 or more times
- You know it will change in the future
- It was tricky to get right

```
theta = arctan2(sqrt(x**2 + y**2),z) * 180/np.pi  
phi = arctan2(y,x) * 180/np.pi + 180
```

- It's a step in a data analysis pipeline

```
data1    = read_in_data(...)  
data2    = read_in_data(...)  
merged   = match_data(data1, data2)  
reduced  = analyze_data(merged)  
plot_data(reduced)
```

# Functions

## Rules of Thumb

- Split up a function if:
  - It has more than ~10 lines (industry)
  - It has more than about **two loops** (science)
  - It takes more than ~5 parameters
  - It does two different things, based on a parameter
- Prefer two, or three simple functions over one big one

# Functions

## Rules of Thumb

- “Magic Functions”
  - It does different things, based on parameters
- Symptoms:
  - 19+ parameters
  - Some parameters can't be used with others
  - matplotlib...

IMO scientists are particularly prone to this

Split it up!

# Functions

## Decorators

`@np.vectorize`

- Put this above your function, and it will work on arrays
- Lots of other use cases for decorators, but this is the one I use on a daily basis

# Functions

## Docstrings

- Explain what your function does!

```
[3]: mf.get_rprof_set_run()
```

```
[ ]:
```

**Signature:** `mf.get_rprof_set_run(run)`

**Docstring:**

Get the radially averaged data sets for some run name.

```
>>> get_rprof_run('M29-768')
```

**File:** `/user/scratch14.4/wthompson/PyPPM/momsfunctions.py`

**Type:** `function`

# Functions: Docstrings

```
def galactocentric_distance(heliocentric_dist, galactic_long, galactic_lat):
```

```
    """\
```

```
    Calculate the galactocentric distance of an object
```

```
    ## INPUT
```

```
    * Heliocentric distance
```

```
    * galactocentric longitude
```

```
    * galactocentric latitude
```

```
    ## OUTPUT
```

```
    * Galactocentric distance (same as input units)
```

```
    Citation: Tobias Westmeier, 2018
```

```
    http://www.atnf.csiro.au/people/Tobias.Westmeier/tools\_coords.php
```

```
    ## EXAMPLES
```

```
    >>> galactocentric_distance(50.0, -57.2, -44.1)
```

```
    43.495214885473814
```

```
    >>> galactocentric_distance(13.0, -5.2, +19.1)
```

```
    11.464307213925464
```

```
    """
```

```
    component_1 = heliocentric_dist * math.cos(galactic_lat) * math.cos(galactic_long) - sun_galaxy_d
```

```
    component_2 = heliocentric_dist * math.cos(galactic_lat) * math.sin(galactic_long)
```

```
    component_3 = heliocentric_dist * math.sin(galactic_lat)
```

```
    galact_dist = (component_1 ** 2 + component_2 ** 2 + component_3)**0.5
```

# Functions

## Doc-tests?

- Explain what your function does!

```
[3]: mf.get_rprof_set_run()
```

```
[ ]:
```

**Signature:** `mf.get_rprof_set_run(run)`

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## Topic 3

# Files & Modules

- Make a text file, give it a **.py** extension
- Congrats, you made a module!

## Topic 3

# Files & Modules

- Make a text file, give it a **.py** extension
- Congrats, you made a module!
- Any python file is a **module** that can be **imported**
- Python looks for modules in your current directory, and PYTHONPATH (that's where e.g. numpy is)
- If you want to find where a module is:
  - Import it, and **print(somemodule) .\_\_file\_\_**

## Topic 3

# Files & Modules

- Importing things should have no side-effects
- At the top level, modules should only include:
  - Constants
  - Functions
  - Classes
- Bad design: importing the module makes a plot show up

## Topic 3

# Packages

- Put a group of modules in a folder
  - Add a file named “\_\_init\_\_.py”
  - Congrats, you have a package!
- 
- Now you can group your modules together, e.g. scipy
  - You could even publish your package to pypi

# Exercises

- Make a function that:
  - Takes three arguments (x,y,z)
  - Returns the distance of the vector
- Make a function that:
  - Takes two points as dictionaries
  - Returns the distance between them
- Put your functions in a module
- Document them!
- Import your module, and use it!
- Extra: Make a function that works over numpy arrays of the vectors (@np.vectorize)

Topic 4

# Using an IDE

## IDEs

# What can an IDE do for you?

- Syntax highlighting
- Multiple cursors
- Integrated terminal
- Warnings as you type (and in panel)
- Autocomplete
- Autoformat
- Code outlines
- Live code sharing, like Google Docs
- Debugging (more on this later)
- Version control (another session?)

Topic 5

# Debugging



# Debugging

## Reading Stack Traces

- <http://cs.franklin.edu/~ansaria/traceback.html>
- Types of Exceptions
  - IndentationError
  - SyntaxError
  - IndexError
  - ValueError
  - TypeError
  - KeyboardInterrupt
- buggy-dictionaries.py

# Debugging

## Using a Debugger

- Breakpoints
- Stepping through
- Stepping in & stepping out
- Conditional breakpoints
- Logpoints
- Hitcounts

Topic 6

# Conda Environments

# Environments

- Reproducible science is important
- Your code should work the same way
  - Next year
  - On someone else's computer
- Installing a new version of a library shouldn't break your other code

# Environments

- Most popular environment system is **virtualenv**
- For our use cases, **conda** is a good choice
  - Comes built in with Anaconda
- Use conda to create a new environment for each project with:
  - Different packages
  - Different versions of python

# Conda Environments

- Create environment.yml
  - `conda env create -f environment.yml`
- List environments:
  - `conda env list`
- Activate an environment:
  - `conda activate <env name>`

# Exercises

- Install an IDE
- Put in your functions from before
- Add a sneaky bug (not a syntax error)
- “Switch seats”, and try out the debugger on each others’ code